- ¹ Title: Tree height and leaf drought tolerance traits shape growth responses across droughts in a temperate
- 2 broadleaf forest

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22 Summary

- As climate change is driving increased drought frequency and severity in many forested regions around
 the world, mechanistic understanding of the factors conferring drought resistance in trees is
 increasingly important. The dendrochronological record provides a window through which we can
 understand how tree size and species' traits shape tree growth responses during droughts.
- We analyzed tree-ring records for twelve species that comprise 97% of the woody productivity of the 25.6-ha ForestGEO plot in a broadleaf deciduous forest of northern Virginia (USA) to test hypotheses on how tree height, microenvironment characteristics, and species' traits shaped drought responses across the three strongest regional droughts over a 60-year period (1950 2009).
- Individual-level drought resistance decreased with tree height, which was strongly correlated with
 exposure to higher evaporative demand and solar radiation. The potentially greater rooting volume of
 larger trees did not confer an advantage in sites with low topographic wetness index. Resistance was
 greater among species whose leaves experienced less shrinkage upon desiccation and lost turgor (wilted)
 at more negative water potentials.
- We conclude that tree height and leaf drought tolerance traits influence growth responses during
 drought, as recorded in the tree-ring record spanning historical droughts. Thus, these factors can be
 useful for predicting future drought responses under climate change.
- Key words: annual growth; crown exposure; drought; Forest Global Earth Observatory (ForestGEO); leaf drought tolerance traits; temperate broadleaf deciduous forest; tree height; tree-ring

41 Introduction

Forests play a critical global role in climate regulation (Bonan, 2008), yet there remains enormous uncertainty as to how the forest-dominated terrestrial carbon sink will respond to climate change (Friedlingstein et al., 2006). An important aspect of this uncertainty lies with physiological responses of trees 44 to drought (Kennedy et al., 2019). In many forested regions around the world, the risk of severe drought is 45 increasing (Trenberth et al., 2014; Dai et al., 2018), often despite increasing precipitation (Intergovernmental Panel on Climate Change, 2015; Cook et al., 2015). Droughts, intensified by climate change, have been 47 affecting forests worldwide and are expected to continue as one of the most important drivers of forest change in the future (Allen et al., 2010, 2015). Understanding forest responses to drought requires elucidation of how tree size, microenvironment, and species' traits jointly influence individual-level drought resistance, and the extent to which their influence is consistent across droughts. Because the resistance and 51 resilience of growth to drought is linked to trees' probability of surviving drought (DeSoto et al. (2020); Liu et al. 2019; DOI: 10.1038/s41558-019-0583-9), understanding growth responses can also help elucidate which trees are most vulnerable to drought-induced mortality. However, it has proven difficult to resolve the many 54 factors affecting tree growth during drought with available forest census data, which only rarely captures extreme drought, and with tree-ring records, which capture multiple droughts but usually only sample a subset of a forest community, typically focusing on a single species or the largest individuals. 57 Many studies have shown that within and across species, large trees tend to be more affected by drought. Greater growth reductions for larger trees were first shown on a global scale by Bennett et al. (2015), and subsequent studies have reinforced this finding (e.g., Hacket-Pain et al., 2016). It has yet to be resolved which of several potential underlying mechanisms most strongly shape these trends in drought response. 61 First, tree height itself may be a primary driver. Taller trees face the biophysical challenge of lifting water greater distances against the effects of gravity and friction (McDowell et al., 2011; McDowell and Allen, 2015; 63 Ryan et al., 2006; Couvreur et al., 2018). Vertical gradients in stem and leaf traits-including smaller and thicker leaves (higher leaf mass per area, LMA), greater resistance to hydraulic dysfunction (i.e., more negative water potential at 50% loss of hydraulic conductivity, more negative P50), and lower hydraulic conductivity at greater heights (Couvreur et al., 2018; Koike et al., 2001; McDowell et al., 2011)—enable trees 67 to become tall (Couvreur et al., 2018). Greater stem capacitance (i.e., water storage capacity) of larger trees may also confer resistance to transient droughts (Phillips et al. 2003; DOI:10.1093/treephys/23.4.237; 69 Scholz et al. 2011). Indeed, tall trees require xylem of greater hydraulic efficiency, such that xylem conduit diameters are wider in the basal portions of taller trees, both within and across species (Olson et al., 2018; Liu et al., 2019), and throughout the conductive systems of angiosperms (Zak et al. 2010, Olson et al. 2014,2018). Wider xylem conduits plausibly make large trees more vulnerable to embolism during 73 drought (Olson et al., 2018), and traits conducive to efficient water transport may also lead to poor ability to recover from or re-route water around embolisms (Roskilly et al., 2019). Larger trees may also have lower drought resistance because of microenvironmental and ecological factors. Their crowns tend to occupy more exposed canopy positions, which are associated with higher evaporative demand (Kunert et al., 2017). Subcanopy trees tend to fare better specifically due to the benefits of a buffered environment (Pretzsch et al., 2018). Counteracting the liabilities associated with tall height, large 79 trees tend to have larger root systems (Enquist and Niklas 2001; DOI: 10.1126/science.1066360), potentially mitigating some of the biophysical challenges they face by allowing greater access to water. Larger root systems—if they grant access to deeper water sources—would be particularly advantageous in drier

microenvironments (e.g., hilltops, as compared to valleys and streambeds) during drought. Finally, tree size-related responses to drought can be modified by species' traits and their distribution across size classes (Meakem et al., 2018; Liu et al., 2019). Understanding the mechanisms driving the greater relative growth reductions of larger trees during drought requires sorting out the interactive effects of height and associated exposure, root water access, and species' traits. Debates have also arisen regarding the traits influencing tree growth responses to drought. Studies within temperate broadleaf forests have observed ring-porous species showing higher drought tolerance than diffuse-porous species (Friedrichs et al., 2009; Elliott et al., 2015; Kannenberg et al., 2019), but this distinction would not hold in the global context (Wheeler et al. 2007, Olson et al. 2020) and does not resolve differences among the many species within each category. Commonly-measured traits including wood density and leaf mass per area (LMA) have been linked to drought responses within some temperate deciduous 93 forests (Abrams, 1990; Guerfel et al., 2009; Hoffmann et al., 2011; Martin-Benito and Pederson, 2015) and across forests worldwide (Greenwood et al., 2017). However, in other cases these traits could not explain drought tolerance (e.g., in a tropical rainforest; Maréchaux et al., 2019), or the direction of response was not always consistent. For instance, higher wood density has been associated with greater drought resistance at a 97 global scale (Greenwood et al., 2017), but correlated negatively with tree performance during drought in a broadleaf deciduous forest in the southeastern United States (Hoffmann et al., 2011). Thus, the perceived influence of these traits on drought resistance may actually reflect indirect correlations with other traits that more directly drive drought responses (Hoffmann et al., 2011). 101 In contrast, hydraulic traits have direct physiological linkages to tree growth and mortality responses to 102 drought. For instance, water potentials at which percent the loss of conductivity surpasses a certain 103 threshold (e.g., P50 and P88, representing 50 and 88% loss of conductivity, respectively) and hydraulic safety 104 margin (i.e., difference between typical minimum water potentials and P50 or P88) correlate with drought 105 performance across global forests (Anderegg et al. 2016). However, these are time-consuming to measure and 106 therefore infeasible for predicting or modeling drought responses in highly diverse forests (e.g., in the tropics). 107 More easily-measurable leaf drought tolerance traits that have direct linkage to plant hydraulic function can 108 explain variation in plant distribution and function (Medeiros et al., 2019). These include leaf area shrinkage 109 upon desiccation (PLA_{dry} ; Scoffoni et al., 2014) and the leaf water potential at turgor loss point (π_{tlp}), i.e., 110 the water potential at which leaf wilting occurs [Bartlett et al. (2016); Zhu et al. 2018]. Both traits correlate 111 with hydraulic vulnerability and drought tolerance as part of unified plant hydraulic systems [Scoffoni et al. (2014); Bartlett et al. (2016); Zhu et al. 2018; but see Farrell et al. 2017]. The abilities of both PLA_{dry} and 113 π_{tlp} to explain tree drought resistance remains untested. 114 Here, we examine how tree height, microenvironment characteristics, and species' traits collectively shape 115 drought resistance, defined as the ratio of annual growth in a drought year to that which would be expected in the absence of drought based on previous years' growth. We test a series of hypotheses and associated 117 specific predictions (Table 1) based on the combination of tree-ring records from the three strongest droughts 118 over a 60-year period (1950 - 2009), species trait measurements, and census and microenvironmental data

interaction with microenvironmental gradients, influences drought resistance. We examine the contemporary

relationship between tree height and microenvironment, including growing season meteorological conditions and crown exposure. We then test whether, consistent with most forests globally, larger-diameter, taller trees

tend to have lower drought resistance in this forest, which is in a region (eastern North America) represented

from a large forest dynamics plot in Virginia, USA. First, we focus on how tree size, alone and in its

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by only two studies in the global review of Bennett et al. (2015). We also test for an influence of potential access to available soil water, which should be greater for larger trees in dry but not in perpetually wet microsites. Finally, we focus on the role of species' traits, testing the hypothesis that species' traits—particularly leaf leaf drought tolerance traits—predict drought resistance. We test predictions that drought resistance is higher in ring-porous than semi-ring and diffuse-porous species and that it is correlated with wood density—either positively (Greenwood et al., 2017) or negatively (Hoffmann et al., 2011) and positively correlated with LMA. We further test predictions that species with low PLA_{dry} have higher drought resistance, and that species whose leaves lose turgor lower water potentials (more negative π_{tlp}) have higher resistance.

134 Materials and Methods

135 Study site and microclimate

Research was conducted at the 25.6-ha ForestGEO (Forest Global Earth Observatory) study plot at the
Smithsonian Conservation Biology Institute (SCBI) in Virginia, USA (38°53'36.6"N, 78°08'43.4"W; Fig. S1)
(Bourg et al., 2013; Anderson-Teixeira et al., 2015a). SCBI is located in the central Appalachian Mountains
near the northern boundary of Shenandoah National Park. Elevations range from 273 to 338 m above sea
level with a topographic relief of 65m (Bourg et al., 2013). Climate is humid temperate, with mean annual
temperature of 12.7°C and precipitation of 1005 mm yr⁻¹ during our study period (1960-2009; source: CRU
TS v.4.01; Harris et al., 2014). Dominant tree taxa within this secondary forest include *Liriodendron*tulipifera, oaks (Quercus spp.), and hickories (Carya spp.; Table 2).

144 Identifying drought years

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We identified the three largest droughts within the time period 1950-2009, defining drought (Slette et al., 2019) as events with anomalously dry peak growing season climatic conditions. Specifically, we used the metric of Palmer Drought Severity Index (PDSI) during May-August (MJJA; Table S1), which were identified by Helcoski et al. (2019) as the months of the current year to which annual tree growth was most sensitive at this site. PDSI divisional data for Northern Virginia were obtained from NOAA (https://www7.ncdc.noaa.gov/CDO/CDODivisionalSelect.jsp) in December 2017. Based on this, we identified the three strongest droughts during the study period (Figs. 1, S1; Table S1).

was preceded by two years of moderate drought during the growing season and severe to extreme drought starting the previous fall. In August 1966, *PDSI* reached its lowest monthly value (-4.82) of the three droughts. The 1977 drought was the least intense throughout the growing season, and it was preceded by 2.5 years of near-normal conditions, making it the mildest of the three droughts. The 1999 drought was preceded by wetter than average conditions until the previous June, but *PDSI* plummeted below -3.0 in October 1998 and remained below this threshold through August 1999.

The droughts differed in intensity and antecedent moisture conditions (Fig. S1, Table S1). The 1966 drought

Data collection and preparation

Within or just outside the ForestGEO plot, we collected data on a suite of variables including tree heights, microenvironment characteristics, and species traits (Table 3). The SCBI ForestGEO plot was censused in 2008, 2013, and 2018 following standard ForestGEO protocols, whereby all free-standing woody stems ≥ 1cm diameter at breast height (DBH) were mapped, tagged, measured at DBH, and identified to species (Condit,

1998). From these census data, we used measurements of DBH from 2008 to calculate historical DBH and data for all stems ≥ 10 cm to analyze functional trait composition relative to tree height (all analyses described below). Census data are available through the ForestGEO data portal (www.forestgeo.si.edu). 166 We analyzed tree-ring data (xylem growth increment) from 571 trees representing the twelve dominant 167 species (Table 2; Fig. S2). Selected species were those with the greatest contributions to woody aboveground 168 net primary productivity $(ANPP_{stem})$ and together comprised 97% of study plot $ANPP_{stem}$ between 2008 and 2013 (Helcoski et al., 2019). Cores (one per tree) were collected within the ForestGEO plot at breast 170 height (1.3m) in 2010-2011 or 2016-2017. In 2010-2011, cores were collected from randomly selected live trees 171 of each species that had at least 30 individuals \geq 10 cm DBH (Bourg et al., 2013). Annual tree mortality censuses were initiated in 2014 (Gonzalez-Akre et al., 2016), and in 2016-2017, cores were collected from all 173 trees found to have died since the previous year's census. We note that drought was probably not a cause of 174 mortality for these trees, as monthly May-Aug PDSI did not drop below -1.75 in these years or the three years prior (2013-2017), and that trees cored dead displayed similar climate sensitivity to trees cored live 176 (Helcoski et al., 2019). Cores were sanded, measured, and crossdated using standard procedures, as detailed 177 in (Helcoski et al., 2019). The resulting chronologies (Fig. 1a) were published in Zenodo (DOI: 10.5281/zenodo.2649302) in association with Helcoski et al. (2019). 179

For each cored tree, we combined tree-ring records and allometric equations of bark thickness to reconstruct DBH for the years 1950-2009. Prior DBH was estimated using the following equation:

$$DBH_Y = DBH_{2008} - 2 * \left[r_{bark,2008} - r_{bark,Y} + \sum_{year=Y}^{2008} r_{ring,Y} \right]$$

Here, Y denotes the year of interest, r_{ring} denotes ring width derived from cores, and r_{bark} denotes bark thickness. Bark thickness was estimated from species-specific allometries based on the bark thickness data

from the site (Anderson-Teixeira et al., 2015b). Specifically, we used linear regression on log-transformed 184 data to relate r_{bark} to diameter inside bark from 2008 data (Table S2), which were then used to determine r_{bark} in the DBH reconstruction. Tree heights (H) were measured by several researchers for a variety of purposes between 2012 and 2019 187 (n=1,518 trees). Methods included direct measurements using a collapsible measurement rod on small trees 188 (NEON, 2018) or a tape measure on recently fallen trees (this study); geometric calculations using clinometer and tape measure (Stovall et al., 2018a) or digital rangefinders (Anderson-Teixeira et al., 2015b; NEON, 2018); and ground-based LiDAR (Stovall et al., 2018b). Rangefinders used either the tangent method 191 (Impulse 200LR, TruPulse 360R) or the sine method (Nikon ForestryPro) for calculating heights. Both methods are associated with some error (Larjavaara and Muller-Landau, 2013), but in this instance there was 193 no clear advantage of one or the other. Measurements from the National Ecological Observatory Network 194 (NEON) were collected near the ForestGEO plot following standard NEON protocol, whereby vegetation of short stature was measured with a collapsible measurement rod, and taller trees with a rangefinder (NEON, 2018). Species-specific height allometries were developed (Table S3) using log-log regression 197 $(\ln[H] \sim \ln[DBH])$. For species with insufficient height data to create reliable species-specific allometries (n=2, JUNI and FRAM), heights were calculated from an equation developed by combining the height 199 measurements across all species. We then used these allometries to estimate H for each drought year, Y, 200 based on reconstructed DBH_Y . The distribution of H across drought years is shown in Fig. S3.

To characterize how environmental conditions vary with height, data were obtained from the NEON tower located <1km from the study area via the neon Utilities package (?). We used wind speed, relative humidity, 203 and air temperature data, all measured over a vertical profile spanning heights from 7.2 m to above the top 204 of the tree canopy (31.0 or 51.8m, depending on censor), for the years 2016-2018 (NEON, 2018). After 205 filtering for missing and outlier values, we determined the daily minima and maxima, which we then aggregated at the monthly scale. 207 Crown position—a categorical variable classifying trees based on exposure to sunlight—was recorded for all 208 cored trees that remained standing during the growing season of 2018 following the protocol of Jennings et al. 209 (1999). Trees were classified as follows: dominant trees were defined as those with crowns above the general 210 level of the canopy, co-dominant trees as those with crowns within the the canopy; intermediate trees as 211 those with crowns below the canopy level, but illuminated from above; and suppressed as those below the 212 canopy and receiving minimal direct illumination from above. Topographic wetness index (TWI), used here as a metric of long-term mean moisture availability, was calculated using the dynatopmodel package in R (Fig. S2) (?). Originally developed by Beven and Kirkby 215 (1979), TWI was part of a hydrological run-off model and has since been used for a number of purposes in 216 hydrology and ecology (Sørensen et al., 2006). TWI calculation depends on an input of a digital elevation 217 model (DEM; ~3.7 m resolution from the elevatr package (?)), and from this yields a quantitative assessment 218 defined by how "wet" an area is, based on areas where run-off is more likely. From our observations in the 219 plot, TWI performed better at categorizing wet areas than the Euclidean distance from the stream. 220 Species' trait data were collected in August 2018 (Tables 2-3; Fig. S4). We sampled small, sun-exposed 221 branches up to eight meters above the ground from three individuals of each species in and around the 222 ForestGEO plot. Sampled branches were re-cut under water at least two nodes above the original cut and re-hydrated overnight in covered buckets under opaque plastic bags before measurements were taken. 224 Rehydrated leaves taken towards the apical end of the branch (n=3 per individual: small, medium, and 225 large) were scanned, weighed, dried at 60° C for ≥ 48 hours, and then re-scanned and weighed. Leaf area was calculated from scanned images using the LeafArea R package (Katabuchi, 2019). LMA was calculated 227 as the ratio of leaf dry mass to fresh area. PLA_{dry} was calculated as the percent loss of area between fresh 228 and dry leaves. Wood density was calculated for ~1cm diameter stem samples (bark and pith removed) as 229 the ratio of dry weight to fresh volume, which was estimated using Archimedes' displacement. We used the 230 rapid determination method of Bartlett et al. (2012) to estimate osmotic potential at turgor loss point (π_{tlp}) . 231 Briefly, two 4 mm diameter leaf discs were cut from each leaf, tightly wrapped in foil, submerged in liquid nitrogen, perforated 10-15 times with a dissection needle, and then measured using a vapour pressure 233 osmometer (VAPRO 5520, Wescor, Logan, UT, USA). Osmotic potential (π_{osm}) given by the osmometer was 234 used to estimate (π_{tlp}) using the equation $\pi_{tlp} = 0.832\pi_{osm}^{-0.631}$ (Bartlett et al., 2012). Statistical Analysis For each drought year, we calculated a metric drought resistance (Rt) as the ratio of basal area increment (BAI; i.e., change in cross-sectional area) during the drought year to the mean BAI over the five years 238 preceding the drought (Lloret et al., 2011). Thus, Rt values <1 and >1 indicate growth reductions and 239 increases, respectively. Because the Rt metric could be biased by directional pre-drought growth trends, we also tried an intervention time series analysis (ARIMA, (?)) that predicted mean drought-year growth based 241 on trends over the preceding ten years and used this value in place of the five-year mean in calculations of

resistance $(Rt_{ARIMA} = \text{observed } BAI/\text{ predicted } BAI)$. The two metrics were strongly correlated (Fig. S5). 243 Visual review of the individual tree-ring sequences with the largest discrepancies between these metrics revealed that Rt was less prone to unreasonable estimates than Rt_{ARIMA} , so we selected Rt as our focal 245 metric, presenting parallel results for Rt_{ARIMA} in the Supplementary Info. In this study we focus exclusively 246 on drought resistance metrics (Rt or Rt_{ARIMA}), and not on the resilience metrics described in Lloret et al. (2011), because (1) we would expect resilience to be controlled by a different set of mechanisms, and (2) the 248 findings of DeSoto et al. (2020) suggest that Rt is a more important drought response metric for 249 angiosperms in that low resistance to moderate droughts was a better predictor of mortality during subsequent severe droughts than the resilience metrics. 251 Analyses focused on testing the predictions presented in Table 1 with Rt as the response variable, and then 252 repeated using Rt_{ARIMA} as the response variable. Models were run for all drought years combined and for 253 each drought year individually. The general statistical model for hypothesis testing was a mixed effects model, implemented in the lme4 package in R (?). In the multi-year model, we included a random effect of 255 tree nested within species and a fixed effect of drought year to represent the combined effects of differences in 256 drought characteristics. Individual year models included a random effect of species. All models included fixed 257 effects of independent variables of interest (Tables 1,3) as specified below. All variables had variance inflation 258 factors <1.5 (1 +/- 0.2). We used AICc to assess model selection, and conditional/marginal R-squared to 259 assess model fit as implemented in the AICcmodavg package in R (?). AICc refers to a corrected version of 260 AICc, and is best suited for small data sizes (see Brewer et al., 2016). 261 To avoid over-fitting models with five species traits (Table 3) across only 12 species, we did not include all 262 traits as fixed effects in a single linear mixed model, but rather conducted individual tests of each species 263 trait to determine the relative importance and appropriateness for inclusion in the main model. These tests 264 followed the model structure specified above, then added ln[H] and ln[TWI] to create a base model against 265 which we tested traits. Trait variables were considered appropriate for inclusion in the main model if they 266 had a consistent direction of response across all droughts and if their addition to the base model improved fit 267 (at $\triangle AICc \ge 1.0$) in at least one drought year (Table S4). We note that we did not use the $\triangle AICc \ge 1.0$ 268 criterion as a test of significance, but rather of whether the variable had enough influence to be considered as 269 a candidate variable in full models. 270 We then determined the top full models for predicting Rt (or Rt_{ARIMA}). To do so, we compared models 271 with all possible combinations of candidate variables, including $ln[H]^*ln[TWI]$ and species traits as specified 272 above. We identified the full set of models within $\Delta AICc=2$ of the best model (that with lowest AICc). When a variable appeared in all of these models and the sign of the coefficient was consistent across models, 274 we viewed this as support for the acceptance/rejection of the associated prediction (Table 1). If the variable 275 appeared in some but not all of these models, and its sign was consistent across models, we considered this partial support/rejection. In presentation of the results below, we note instances where the Rt_{ARIMA} model 277 disagreed with the Rt model, but otherwise do not discuss the Rt_{ARIMA} model. 278 All analysis beyond basic data collection was performed using R version 3.6.2 (R Core Team, 2020). Other 279 R-packages used in analyses are listed in the Supplementary Information (Appendix S1). All data, code, and results are available through the SCBI-ForestGEO organization on GitHub 281 (https://github.com/SCBI-ForestGEO: SCBI-ForestGEO-Data and McGregor climate-sensitivity-variation 282 repositories), with static versions corresponding to data and analyses presented here archived in Zenodo 283 (DOIs: 10.5281/zenodo.3604993 and [TBD], respectively.

285 Results

microhabitats.

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Tree height and microenvironment 286 In the years for which we have vertical profiles in climate data (2016-2018), taller trees—or those in dominant 287 crown positions—were generally exposed to higher evaporative demand during the peak growing season 288 months (May-August; Fig. 2). Specifically, maximum daily wind speeds were significantly higher above the 289 top of the canopy (40-50m) than within and below (10-30m) (Fig. 2a). Relative humidity was also somewhat 290 lower during June-August, ranging from ~50-80% above the canopy and ~60-90% in the understory (Fig. 2b). 291 Air temperature did not vary consistently across the vertical profile (Fig. 2c). 292 Crown position varied as expected with height (dominant > co-dominant > intermediate > suppressed), but 293 with substantial variation (Fig. 2d). There were significant differences in height across all crown position 294 classes (Fig. 2d). A comparison test between height and crown position data from the most recent 295 ForestGEO census (2018) revealed a correlation of 0.73. 296 Community-level drought responses 297 At the community level, cored trees showed substantial growth reductions in all three droughts, with a mean 298 Rt of 0.86 in 1966 and 1999, and 0.84 in 1977 (Fig. 2b). Across the entire study period (1950-2009), the 299 focal drought years were the three years with the largest fraction of trees exhibiting $Rt \leq 0.7$. Specifically, in each drought, roughly 30% of the cored trees had growth reductions of > 30% (Rt < 0.7): 29% in 1966, 32% 301 in 1977, and 27% in 1999. However, some individuals exhibited increased growth, i.e., Rt > 1.0: 26% of trees 302 in 1966, 22% in 1977, and 26% in 1999. 303 In the context of the multivariate model, Rt did not vary across drought years. That is, drought year as a 304 variable did not appear in any of the top models -i.e., models that were statistically indistinguishable 305 $(\Delta AICc < 2)$ from the best model. Tree height, microenvironment, and drought resistance Taller trees (based on H in the drought year) showed stronger growth reductions during drought (Table 1; 308 Figs. 4, S6). Specifically, ln[H] appeared, with a negative coefficient, in the best model (($\Delta AICc=0$) and all 309 top models when evaluating the three drought years together (Tables S6-S7). The same held true for 1966 310 individually. For the 1977 drought, ln[H] did not appear in the best model, but was included, with a 311 negative coefficient, among the top models -i.e., models that were statistically indistinguishable ($\Delta AICc<2$) 312 from the best model (Tables 1, S6-S7). For the 1999 drought, ln[H] had no significant effect. 313 Rt had a significantly negative response to ln[TWI] across all drought years combined (Figs. 4, S6, Table 314 S6-S7). The effect was also significant for 1977 and 1999 individually (Fig. 4, Table S6). When Rt_{ARIMA} 315 was used as the response variable, the effect was significant in 1977, and included in some of the top models 316 in 1966 and 1999 (Table S7). This negates the idea that trees in moist microsites would be less affected by 317 drought. Nevertheless, we tested for a ln[H] * ln[TWI] interaction, a negative sign of which could indicate 318 that smaller trees (presumably with smaller rooting volume) are more susceptible to drought in drier 319 microenvironments with a deeper water table. This hypothesis was rejected, as the ln[H] * ln[TWI]interaction was never significant, and had a positive sign in any top models in which it appeared (Tables 1, 321 S6-S7). This term did appear with a positive coefficient in the best Rt_{ARIMA} model for all years combined 322 (Table S7), indicating that the negative effect of height on Rt was significantly stronger in wetter

Species, as a factor in ANOVA, had significant influence (p<0.05) on all traits (wood density, LMA, 326 PLA_{dry} , and π_{tlp}), with more significant pairwise differences for wood density and PLA_{dry} than for LMA327 and π_{tlp} (Table 2, Fig. S4). Drought resistance also varied across species, overall and in each drought year 328 (Fig. 3). Averaged across all droughts, Rt was lowest in Liriodendron tulipifera (mean Rt = 0.66) and 329 highest in Fagus grandifolia (mean Rt = 0.99). Wood density, LMA, and xylem porosity were all poor predictors of Rt (Tables 1,S4-S5). Wood density and 331 LMA were never significantly associated with Rt in the single-variable tests and were therefore excluded 332 from the full models. Xylem porosity was also excluded from the full models, as it had no significant 333 influence for all droughts combined and had contrasting effects in the individual droughts: whereas 334 ring-porous species had higher Rt than diffuse- and semi-ring- porous species in the 1966 and 1999 droughts, 335 they had lower Rt in 1977 (Table S4). It is noteworthy that the two diffuse-porous species in our study, 336 Liriodendron tulipifera and Fagus grandifolia, were at opposite ends of the Rt spectrum (Fig. 3), further refuting the idea that xylem porosity is a useful predictor of Rt in the context of this study. 338 In contrast, PLA_{dry} , and π_{tlp} were both negatively correlated to drought resistance (Figs. 4, S6; Tables 339 1,S4-S7). Both had consistent signs across all droughts, and their inclusion at least marginally improved the model ($\Delta AICc > 1.0$) for at least one of the three droughts (Table S4), qualifying them as candidate 341 variables for the full model. PLA_{dry} had a significant influence, with negative coefficient, in full models for 342 the three droughts combined and for the 1966 drought individually (Fig. 4; Tables S6-S7). For 1977 and 1999, it was included with a negative coefficient in some of the top models (Tables S6-S7). π_{tlp} was included with a negative coefficient in the best model for both all droughts combined and for the 1977 drought 345 individually (Fig. 4; Table S6). It was also included in some of the top models for 1999 (Tables S6-S7).

Discussion

Species' traits and drought resistance

Tree height, microenvironment, and leaf drought tolerance traits shaped tree growth responses across three droughts at our study site (Table 1, Fig. 4). The greater susceptibility of larger trees to drought, similar to 349 forests worldwide (Bennett et al., 2015), was driven primarily by their height (Stovall et al., 2019). Taller 350 height was likely a liability in itself, and was also associated with greater exposure to conditions that would 351 promote water loss and heat damage during drought (Fig. 2). There was no evidence that greater 352 availability of, or access to, soil water availability increased drought resistance; in contrast, trees in wetter 353 topographic positions had lower Rt (Zuleta et al., 2017; Stovall et al., 2019), and the larger potential rooting volume of large trees provided no advantage in the drier microenvironments. The negative effect of height on 355 Rt held after accounting for species' traits, which is consistent with recent work finding height had a stronger 356 influence on mortality risk than forest type during drought (Stovall et al. 2020). Drought resistance was not 357 consistently linked to species' LMA, wood density, or xylem type (ring- vs. diffuse porous), but was 358 negatively correlated with leaf drought tolerance traits (PLA_{dry}, π_{tlp}) . This is the first study to our 359 knowledge linking PLA_{dry} and π_{tlp} to growth reduction during drought. The directions of these responses were consistent across droughts (Table S6), supporting the premise that they were driven by fundamental 361 physiological mechanisms. However, the strengths of each predictor varied across droughts (Fig. 4; Tables 362 S6-S7), indicating that drought characteristics interact with tree size, microenvironment, and traits to shape which individuals are most affected. These findings advance our knowledge of the factors that make trees 364 vulnerable to growth declines during drought and, by extension, likely make them more vulnerable to

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mortality (Sapes et al., 2019).
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    The droughts considered here were of a magnitude that has occurred with an average frequency of
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    approximately once every 10-15 years (Fig. 1a, Helcoski et al., 2019) and had substantial but not devastating
    impacts on tree growth (Figs. 1b). These droughts were classified as severe (PDSI < -3.0; 1977) or extreme
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    (PDSI < -4.0; 1966, 1999) at our site and have been linked to tree mortality in the eastern United States
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    (Druckenbrod et al., 2019). However, extreme, multiannual droughts such as the so-called "megadroughts" of
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    this type that have triggered massive tree die-off in other regions (e.g., Allen et al., 2010; Stovall et al., 2019)
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    have not occurred in the Eastern United States within the past several decades (Clark et al., 2016). Of the
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    droughts considered here, the 1966 drought, which was preceded by two years of dry conditions (Fig. S1),
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    severely stressed a larger portion of trees (Fig. 1b). The tendency for large trees to have lowest resistance
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    was most pronounced in this drought, consistent with other findings that this physiological response increases
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    with drought severity (Bennett et al., 2015; Stovall et al., 2019). Across all three droughts, the majority of
    trees experienced reduced growth, but a substantial portion had increased growth (Fig. 1b), potentially due
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    to decreased leaf area of competitors during the drought (REF-if we can find one), and consistent with
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    prior observations that smaller trees can exhibit increased growth rates during drought (Bennett et al., 2015).
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    It is likely because of the moderate impact of these droughts, along with other factors influencing tree growth
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    (e.g., stand dynamics), that our best models characterize only a modest amount of variation in Rt: 11-12%
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    for all droughts combined, and 18-25% for each individual drought (Fig. S6; Table S6).
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    Consistent with studies in other forests worldwide (Bennett et al., 2015), taller trees in this forest exhibited
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    lower drought resistance. Mechanistically, this is consistent with, and reinforces, previous findings that
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    biophysical constraints make it impossible for trees to efficiently transport water to great heights and
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    simultaneously maintain strong resistance and resilience to drought-induced embolism (Olson et al., 2018;
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    Couvreur et al., 2018; Roskilly et al., 2019). Taller trees also face dramatically different microenvironments
388
    (Fig. 2). They are exposed to higher wind speeds and lower humidity (Fig. 2a-b), resulting in higher
389
    evaporative demand. Unlike other temperate forests where modestly cooler understory conditions have been
    documented (Zellweger et al. 2019), particularly under drier conditions (Davis et al. 2019), we observed no
391
    significant variation in air temperatures across the vertical profile (Fig. 2c). More critically for tree
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    physiology, leaf temperatures can become significantly elevated over air temperature under conditions of high
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    solar radiation and low stomatal conductance (Campbell & Norman; Rey-Sanchez et al. 2016). Under
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    drought, when air temperatures tend to be warmer, direct solar radiation tends to be higher (because of less
    cloud cover), and less water is available for evaporative cooling of the leaves, trees with sun-exposed crowns
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    may not be able to simultaneously maintain leaf temperatures below damaging extremes and avoid
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    drought-induced embolism. Indeed, previous studies have shown lower drought resistance in more exposed
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    trees (Liu and Muller, 1993; Suarez et al., 2004; Scharnweber et al., 2019). Unfortunately, collinearity
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    between height and crown exposure in this study (Fig. 2d) makes it impossible to confidently partition
400
    causality. Additional research comparing drought responses of early successional and mature forest stands,
    along with short and tall isolated trees, would be valuable for more clearly disentangling the roles of tree
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    height and crown exposure.
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    Belowground, taller trees would tend to have larger root systems (Enquist and Niklas 2001; DOI:
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    10.1126/science.1066360), but this does not necessarily imply that they have greater access to or reliance on
    deep soil-water resources that may be critical during drought. Rather, larger trees may allocate more to
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    abundant shallow roots that are beneficial for taking up water from rainstorms (Meinzer et al. 1999; DOI:
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10.1007/s004420050931). In any case, the potentially greater access to water did not override the
408
    disadvantage conferred by height-and, in fact, greater moisture access in non-drought years (here, higher
    TWI) appears to make trees more sensitive to drought (Zuleta et al., 2017; Stovall et al., 2019). This may be
410
    because moister habitats would tend to support species and individuals with more mesophytic traits (Bartlett
411
    et al. 2016 Drought. DOI: 10.1890/15-0468.1; Mencuccini 2003, DOI: 10.1046/j.1365-3040.2003.00991.x)
    (Medeiros et al., 2019), potentially growing to greater heights (e.g., Detto et al. DOI:
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    10.1371/journal.pone.0076296), and these are then more vulnerable when drought hits. The observed
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    height-sensitivity of Rt, together with the lack of conferred advantage to large stature in drier topographic
    positions, agrees with the concept that physiological limitations to transpiration under drought shift from
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    soil water availability to the plant-atmosphere interface as forests age (Bretfeld et al., 2018), such that tall,
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    dominant trees are the most sensitive in mature forests. Again, additional research comparing drought
    responses across forests with different tree heights and water availability would be valuable for disentangling
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    the relative importance of above- and belowground mechanisms across trees of different size.
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    The development of tree-ring chronologies for the twelve most dominant tree species at our site (Helcoski
421
    et al., 2019; Bourg et al., 2013) gave us the sample size to compare historical drought responses across
422
    species (Fig. 3) and associated traits at a single site (see also Elliott et al., 2015). Our study reinforced
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    current understanding (see Introduction) that wood density and LMA are not reliably linked to drought
424
    resistance (Table 1). Contrary to previous studies in temperate deciduous forests, we did not find an
425
    association between xylem porosity and drought tolerance, as the two diffuse-porous species, Liriodendron
426
    tulipifera and Fagus grandifolia, were at opposite ends of the Rt spectrum (Fig. 3). While the low Rt of L.
427
    tulipifera is consistent with other studies (Elliott et al., 2015), the high Rt of F. grandifolia contrasts with
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    studies identifying diffuse porous species in general (Elliott et al., 2015; Kannenberg et al., 2019), and the
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    genus Faqus in particular (Friedrichs et al., 2009), as drought sensitive. There are two potential explanations
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    for this discrepancy. First, other traits can and do override the influence of xylem porosity on drought
431
    resistance. Ring-porous species are restricted mainly to temperate deciduous forests (Wheeler et al. 2007),
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    while highly drought-tolerant diffuse-porous species exist in other biomes (REFS). Fagus grandifolia had
433
    intermediate \pi_{tlp} and low PLA_{dry} (Fig. S4), which would have contributed to its drought resistance (Fig. 4;
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    see discussion below). A second explanation of why F. grandifolia trees at this particular site had higher Rt
435
    is that the sampled individuals, reflective of the population within the plot, are generally shorter and in
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    less-dominant canopy positions compared to most other species (Fig. S4). The species, which is highly
437
    shade-tolerant, also has deep crowns (Anderson-Teixeira et al., 2015b), implying that a lower proportion of
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    leaves would be affected by harsher microclimatic conditions at the top of the canopy under drought (Fig. 2).
439
    Thus, the high Rt of the sampled F. grandifolia population can be explained by a combination of fairly
440
    drought-resistant leaf traits, shorter stature, and a buffered microenvironment.
441
    Concerted measurement of tree-rings and leaf drought tolerance traits of emerging importance (Scoffoni
442
    et al., 2014; Bartlett et al., 2016; Medeiros et al., 2019) allowed novel insights into the role of drought
    tolerance traits in shaping drought response. The finding that PLA_{dry} and \pi_{tlp} can be useful for predicting
444
    drought responses of tree growth (Fig. 4; Table 1) is both novel and consistent with previous studies linking
445
    these traits to habitat and drought tolerance. Previous studies have demonstrated that \pi_{tlp} and PLA_{dry} are
    physiologically meaningful traits linked to species distribution along moisture gradients [Maréchaux et al.
447
    (2015); Fletcher et al. (2018); Medeiros et al. (2019); Simeone et al. (2019); Rosas et al. (2019); Zhu et
    al. 2018], and our findings indicate that these traits also influence drought responses. Furthermore, the
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drought-intolerant than drought-tolerant plant functional type. Further, it adds support to the idea that this trait is useful for categorizing and representing species' drought responses in models (Powell et al., 2017). 452 Because both PLA_{dry} and π_{tlp} can be measured relatively easily (Bartlett et al., 2012; Scoffoni et al., 2014), 453 they hold promise for predicting drought growth responses across diverse forests. The importance of predicting drought responses from species traits increases with tree species diversity; whereas it is feasible to 455 study drought responses for all dominant species in most boreal and temperate forests (e.g., this study), this 456 becomes difficult to impossible for species that do not form annual rings, and for diverse tropical forests. 457 Although progress is being made for the tropics (Schöngart et al., 2017), a full linkage of drought tolerance 458 traits to drought responses would be invaluable for forecasting how little-known species and whole forests will 459 respond to future droughts (Christoffersen et al. 2016) (Powell et al., 2017). As climate change drives increasing drought in many of the world's forests (Trenberth et al., 2014; Intergovernmental Panel on Climate Change, 2015), the fate of forests and their climate feedbacks will be 462 shaped by the biophysical and physiological drivers observed here. Our results, consistent with other 463 observations around the world, imply that the tallest, most exposed trees will be most affected (Bennett 464 et al., 2015; Stovall et al., 2019). We show that, at least within the mature forest studied here, the 465 vulnerability conferred by tall height and associated crown exposure outweigh any advantage of a larger root 466 system, even in drier microenvironments. This would suggest that the drought responses of trees in mature 467 forests are more strongly differentiated along the size spectrum by their above- than below-ground 468 environment. The same may not be true of systems where short trees exist outside of a buffered understory 469 environment-i.e., open grown trees or short-statured, early-successional forests. The latter appear to be 470 limited more strongly by root water access during drought (Bretfeld et al., 2018), and would also be 471 dominated by species with different traits. The earlier-successional species at our site (Liriodendron 472 tulipifera, Quercus spp., Fraxinus americana) display a mix of traits conferring drought tolerance and 473 resistance (Table 2), while the late-successional Fagus grandifolia displayed high drought resistance, in part 474 because it exists primarily within a buffered microenvironment. Further research on how leaf drought 475 tolerance traits and drought vulnerability change over the course of succession would be valuable for addressing how drought tolerance changes as forests age (e.g. Rodríguez-Catón et al., 2015). In the 477 meantime, the results of this study advance our knowledge of the factors conferring drought resistance in a mature forest, opening the door for more accurate forecasting of forest responses to future drought. 479

observed linkage of π_{tlp} to Rt in this forest aligns with observations in the Amazon that π_{tlp} is higher in

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491 Author Contribution

- 492 KAT, IM, and AJT designed the research. Tree-ring chronologies were developed by RH under guidance of
- ⁴⁹³ AJT and NP. Trait data were collected by IM, JZ under guidance of NK and LS. Other plot data were
- collected by IM, AS, EGA, and NB under guidance of EGA and WM. Data analyses were performed by IM
- under guidance of KAT and VH. KAT and IM interpreted the results. IM and KAT wrote the first draft of
- manuscript, and all authors contributed to revisions.

497 Supplementary Information

- 498 redo this list!!
- Table S1: Species-specific bark thickness regression equations
- Table S2: Species-specific height regression equations
- Table S3: Palmer drought severity index (PDSI) by month for focal droughts
- Figure S1: Map of ForestGEO plot showing TWI and location of cored trees
- Figure S2: Time series of Palmer Drought Severity Index (PDSI) for the 2.5 years prior to each focal drought
- $_{504}$ Figure S3: Height (from reconstructed DBH) by crown position across the three focal droughts and in the
- year of measurement (2018)

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